Week 2



Case Study - Understanding the data

- Descriptive statistics
 - Univariate statistics
 - Pairwise correlations
 - Histogram

$$f(a) + \sum_{k=1}^{n} \frac{1}{k!} \frac{d^{k}}{dt^{k}} \bigg|_{t=0} f(u(t)) + \int_{0}^{1} \frac{(1-t)^{n}}{n!} \frac{d^{n+1}}{dt^{n+1}} f(u(t)) dt.$$

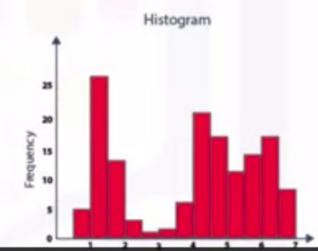
 $F_{X,Y}(x,y)$ satisfies

$$F_{X,Y}(x,y) = F_X(x)F_Y(y),$$

or equivalently, their joint density $f_{X,Y}(x,y)$ satisfies

$$f_{X,Y}(x,y) = f_X(x)f_Y(y).$$

Histograms are a good way to understand how values or a variable are distributed, and what sorts of data preparation may be needed to make the variable more useful in a model.



First, these statistics included Hearst, univariates, and statistics on each

variable, such as mean,





Case study - Looking at data quality

- Data quality
 - Missing values
 - Invalid or misleading values



From the information provided, certain values can be re-coded or perhaps even dropped if



Question

The Data Understanding stage encompasses sorting the data.

- O True.
- False.

Correct
Correct.

Skip

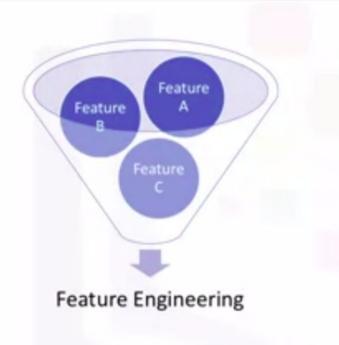
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Examples of data cleansing

	A	В	C	D	E
	Name	Date	Age	Location	Country
!	John Doe	2012 02 20	32	ON	CAN
ı	May Lag	2013 02 33	2	ON	CA
ļ	Henry Oon	30-Sep-12	35	Ontario	CANADA
,	Kelly, Tom	2015 02 20	65	ON	CA
5	John Kell	2016 02 20		AB	CA
7	Henry Oon	30-Sep-12	35	Ontario	CANADA
8	Invalid Values				
	Invalid values				
	Missing Data				
	Remove Duplicates				
	Formatting				

or invalid values and removes duplicates, toward ensuring that everything is

Using domain knowledge

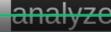


Feature engineering is the process of using domain knowledge of the data to create features that make the machine learning algorithms work.

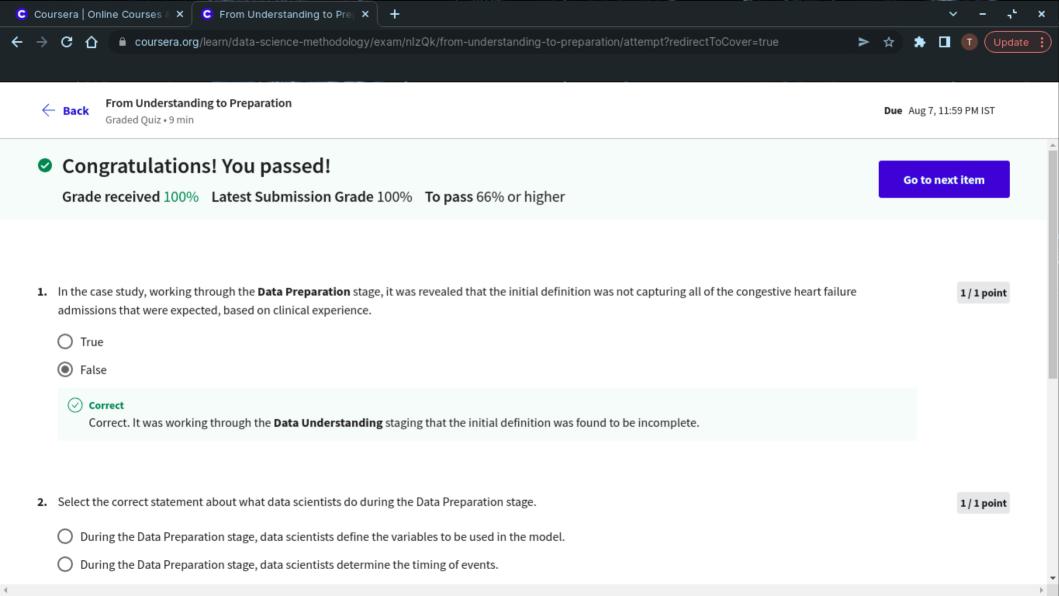
Feature engineering is critical when machine learning tools are being applied to analyze the data.

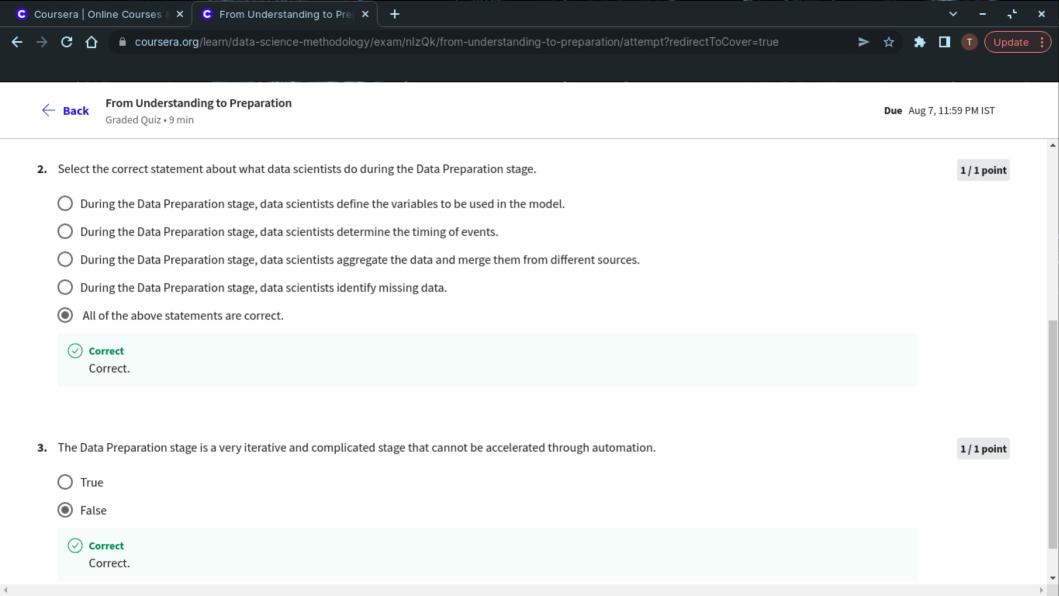
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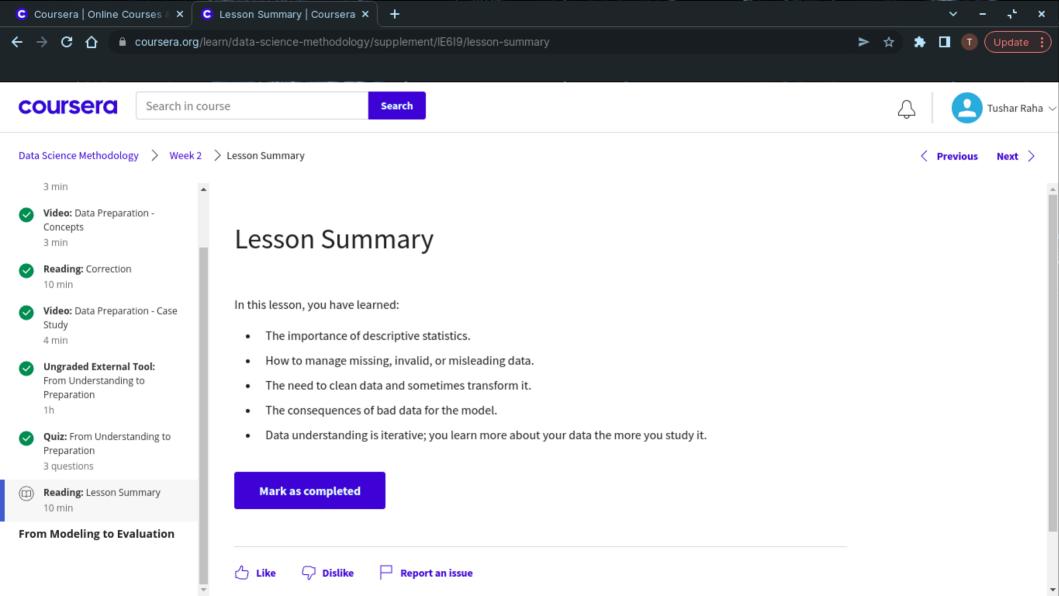




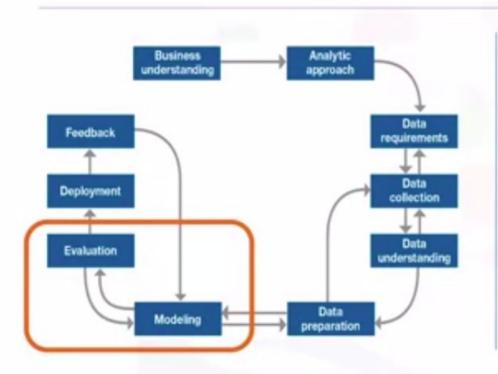








From Modeling to Evaluation



Modeling

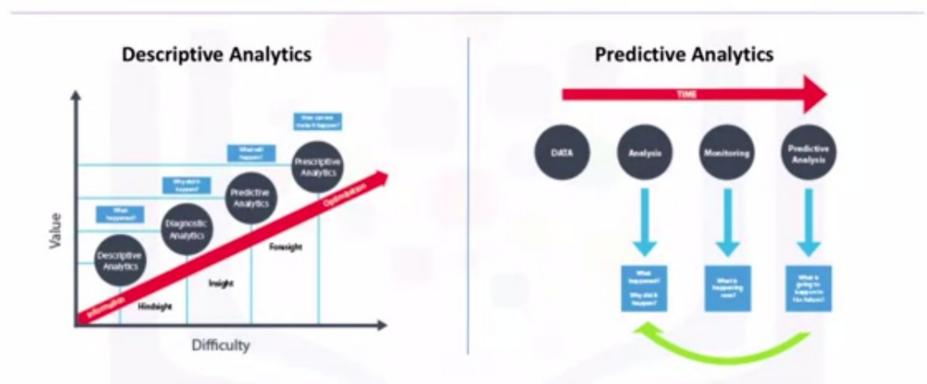
 In what way can the data be visualized to get to the answer that is required?



Evaluation

 Does the model used really answer the initial question or does it need to be adjusted?

Data Modeling – Using Predictive or Descriptive?

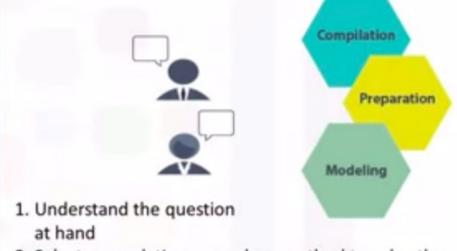


An example of a descriptive model might examine things like: if a person did

this,

Understanding the question





- Select an analytic approach or method to solve the problem
- 3. Obtain, understand, prepare, and model the data

Question

A training set is used to build a predictive model.

- True.
- O False.



✓ Correct

Correct.

Skip

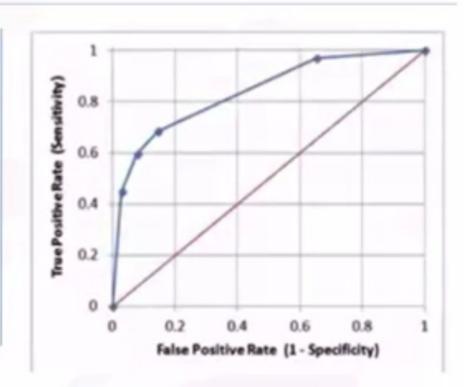
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Case Study – Using the ROC curve

Diagnostic tool for classification model evaluation

- Classification model performance
- True-Positive Rate vs False-Positive Rate
- Optimal model at maximum separation



which was first developed during World War II to detect enemy aircraft on radar.

Question

Model evaluation can have two main phases: a diagnostic measures phase and statistical significance testing.

Diag

all True.

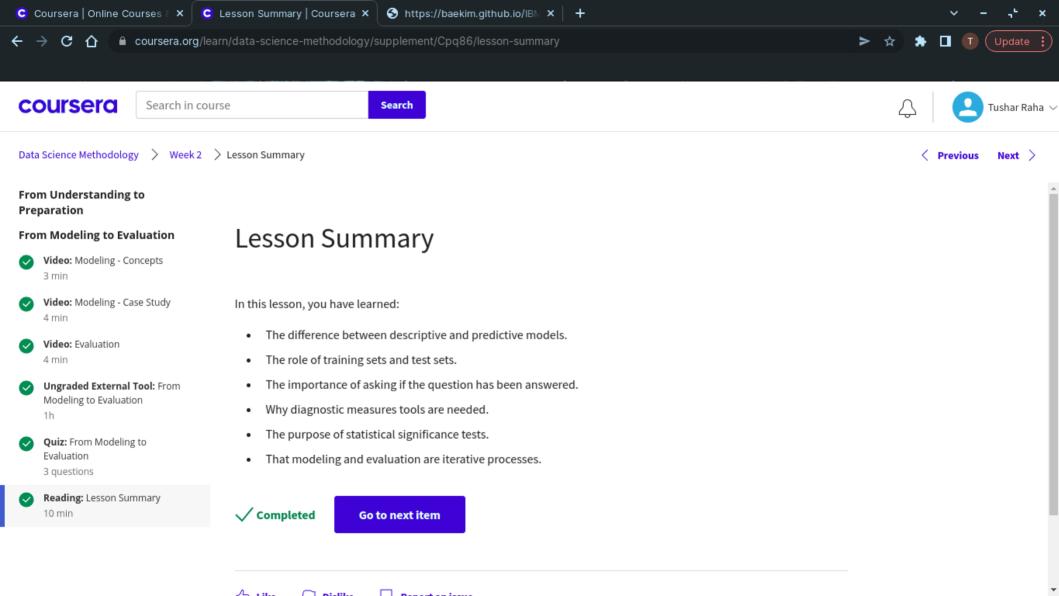
✓ Correct

O False.

Correct.

Skip

Continue



Week 3



Case Study – Understand the results

Assimilate knowledge for business

- Practical understanding of the meaning of model results
- Implications of model results for designing intervention actions



In preparation for solution deployment, the next step was to assimilate the knowledge



Case Study – Gathering application requirements

Application requirements

- Automated, near-real-time risk assessments of CHF inpatients
- Easy to use
- Automated data preparation and scoring
- Up-to-date risk assessment to help clinicians target high-risk patients







From Deployment to Feedback



Once the model is evaluated and the data scientist is confident it will work, it is deployed and put to the ultimate test

Actual real-time use in the field

Once the model is evaluated and the data scientist is confident it'll work, it is



Case Study – Assessing model performance

Define review process

- To measure results of applying the risk model to the CHF patient population
- Track patients who received intervention
 - Actual readmission outcomes
- Measure effectiveness of intervention
 - Compare readmission rates before & after model implementation

Domain Expertise Data Quality Time Accuracy

for measuring the results of a "flying to risk" model of the congestive heart

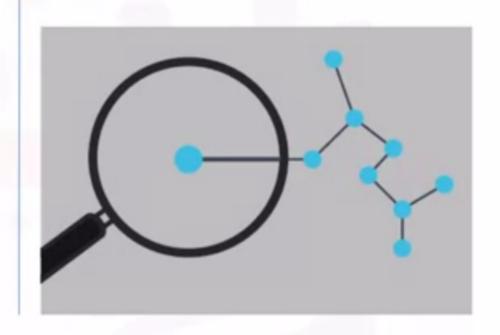
failure



Case Study – Refinement

Refine model

- Initial review after the first year of implementation
- Based on feedback data and knowledge gained
- Participation in intervention program
- Possibly incorporate detailed pharmaceutical data originally deferred
- Other possible refinements as yet unknown

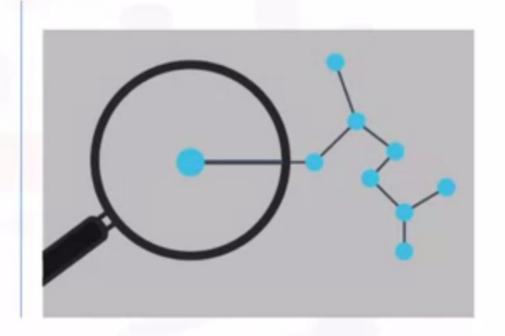




Case Study – Refinement

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Question

Ca

The data science methodology is highly iterative, ensuring the refinement at each stage in the game.

Revi

O False.

True.

Rede

fe th Correct

Correct.

Skip

Continue

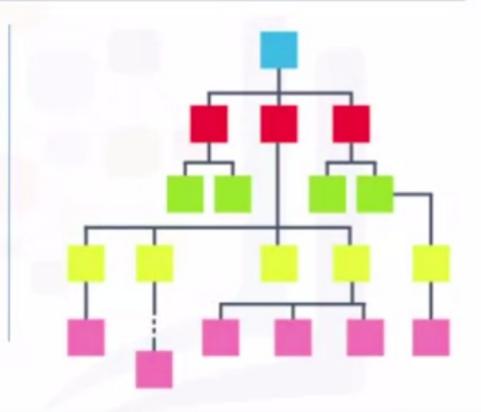


Case Study - Redeployment

Review and refine intervention actions

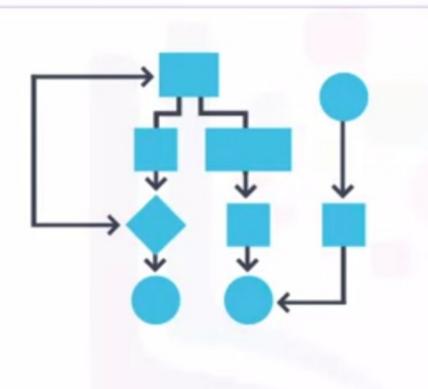
Redeploy

 Continue modeling, deployment, feedback, and refinement throughout the life of the intervention program





From problem to approach



Thinking like a data scientist!

- Forming a concrete business or research problem,
- Collecting and analyzing data,
- Building a model, and
- Understanding the feedback after model deployment.

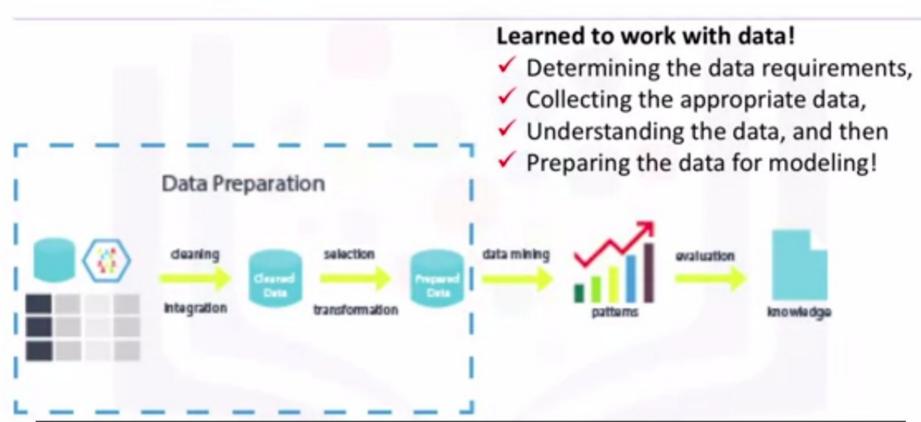
Learned importance of:

- Understanding the question, and
- Picking the most effective analytic approach

forming a concrete business or research problem, collecting and analyzing data,

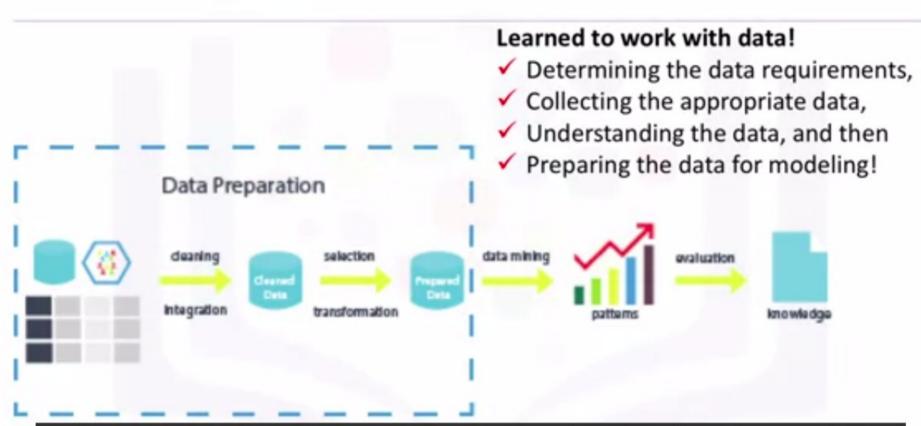


To working with the data



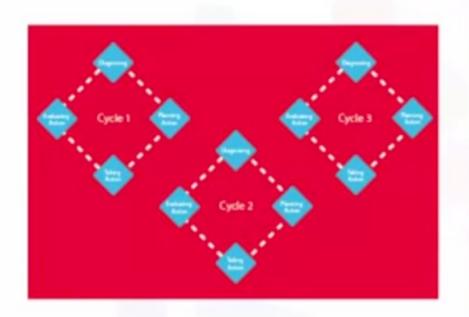
understanding the data, and then preparing the data for modeling!

To working with the data



understanding the data, and then preparing the data for modeling!

To deriving the answer



Once the analytic approach is selected, learned how to derive the answer:

- Evaluating and deploying the model,
- Getting feedback on it, and
- Using that feedback constructively so as to improve the model.

Remember that the stages of this methodology are *iterative*!

You've also learned how to model the data by using the appropriate analytic approach, SKILLS NETWORK

The methodology in a nutshell...

The **Data Science Methodology** aims to answer the following 10 questions in this prescribed sequence:

From problem to approach:

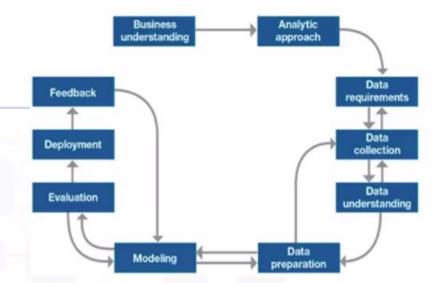
- 1. What is the problem that you are trying to solve?
- 2. How can you use data to answer the question?

Working with the data:

- 3. What data do you need to answer the question?
- 4. Where is the data coming from (identify all sources) and how will you get it?
- 5. Is the data that you collected representative of the problem to be solved?
- 6. What additional work is required to manipulate and work with the data?

Deriving the answer:

- 7. In what way can the data be visualized to get to the answer that is required?
- 8. Does the model used really answer the initial question or does it need to be adjusted?
- 9. Can you put the model into practice?
- 10. Can you get constructive feedback into answering the question?



tools, at the right time, in the right order, to the address the right problem

) 2:57 / 3:30

